

# **Quality Adjustment and Hedonics: A Unified Approach**

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# Introduction

- The paper takes a consumer demand perspective to the problem of adjusting product prices for quality change.
- The various approaches to the problem of **quality adjustment** can be seen as special cases of the general framework.
- The special cases include
  - (i) the use of **inflation adjusted carry forward and carry backward prices,**
  - (ii) the use of **hedonic regressions** and
  - (iii) the **estimation of Hicksian reservation prices.**
- I will not be able to cover the last topic in this presentation and some hedonic regression topics will not be covered in detail.
- A Table of Contents for the paper will appear on the following slides.

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# The Basic Consumer Theory Framework

- **Notation:** Let  $\mathbf{p}^t \equiv [p_{t1}, \dots, p_{tN}]$  and  $\mathbf{q}^t \equiv [q_{t1}, \dots, q_{tN}]$  denote the price and quantity vectors for time periods  $t = 1, \dots, T$ .
- The period  $t$  quantity for product  $n$ ,  $q_{tn}$ , is equal to **total purchases** of product  $n$  by purchasers or to the sales of product  $n$  by the outlet (or group of outlets) for period  $t$ , while the period  $t$  price for product  $n$ ,  $p_{tn}$ , is equal to the **value of sales** (or **purchases**) of product  $n$  in period  $t$ ,  $v_{tn}$ , divided by the corresponding total quantity sold (or purchased),  $q_{tn}$ .
- Thus  $p_{tn} \equiv v_{tn}/q_{tn}$  is the **unit value price** for product  $n$  in period  $t$  for  $t = 1, \dots, T$  and  $n = 1, \dots, N$ .
- Initially, we assume that all prices, quantities and values are positive; in subsequent sections, this assumption will be relaxed.
- I have in mind a **scanner data context** for an elementary category.

## The Basic Consumer Theory Framework (cont)

- Let  $q \equiv [q_1, \dots, q_N]$  be a generic quantity vector.
- In order to compare various methods for comparing the value of alternative combinations of the  $N$  products, it is necessary that a *valuation function* or *aggregator function*,  $Q(q)$ , exist.
- This function allows us to value alternative combinations of products; if  $Q(q^2) > Q(q^1)$ , then purchasers of the products place a higher utility value on the vector of purchases  $q^2$  than they place on the vector of purchases  $q^1$ .
- The function  $Q(q)$  can also act as an *aggregate quantity level* for the vector of purchases,  $q$ .
- Thus  $Q(q^t)$  can be interpreted as an *aggregate quantity level* for the period  $t$  vector of purchases,  $q^t$ , and the ratios,  $Q(q^t)/Q(q^1)$ ,  $t = 1, \dots, T$ , can be interpreted as *fixed base quantity indexes* covering periods 1 to  $T$ .

## Properties of $Q(q^t)$

- $Q(q)$  has the following properties:
  - (i)  $Q(q) > 0$  if  $q \gg 0_N$ ;
  - (ii)  $Q(q)$  is **nondecreasing** in its components;
  - (iii)  $Q(\lambda q) = \lambda Q(q)$  for  $q \geq 0_N$  and  $\lambda \geq 0$ ; (**linear homogeneity**);
  - (iv)  $Q(q)$  is a **continuous concave function** over the nonnegative orthant.
- Assumption (iii), linear homogeneity of  $Q(q)$ , is a somewhat restrictive assumption.
- However, this assumption is required to ensure that the **aggregate price level**,  $P(p,q) \equiv p \cdot q / Q(q)$  that corresponds to  $Q(q)$  does not depend on the scale of  $q$ .
- Property (iv) will ensure that the first order necessary conditions for the budget constrained maximization of  $Q(q)$  are also sufficient.

## The Aggregate Price Level Defined

- Let  $p \equiv [p_1, \dots, p_N] > 0_N$  and  $q \equiv [q_1, \dots, q_N] > 0_N$  be generic price and quantity vectors with  $p \cdot q \equiv \sum_{n=1}^N p_n q_n > 0$ .
- Then the *aggregate price level*,  $P(p, q)$  that corresponds to the *aggregate quantity level*  $Q(q)$  is defined as follows:

(1)  $P(p, q) \equiv p \cdot q / Q(q)$ .

- Thus the implicit price level that is generated by the generic price and quantity vectors,  $p$  and  $q$ , is equal to the value of purchases,  $p \cdot q$ , deflated by the aggregate quantity level,  $Q(q)$ .
- Note that using these definitions, the product of the aggregate price and quantity levels equals the value of purchases during the period,  $p \cdot q$ . (**Product Test for Levels**).

## More Introduction

- Once the **functional form** for the aggregator function  $Q(q)$  is **known**, then the *aggregate quantity level for period  $t$* ,  $Q^t$ , can be calculated in the obvious manner:

$$(2) Q^t \equiv Q(q^t); \quad t = 1, \dots, T.$$

- Using definition (1), the corresponding **period  $t$  aggregate price level**,  $P^t$ , can be calculated as follows:

$$(3) P^t \equiv p^t \cdot q^t / Q(q^t); \quad t = 1, \dots, T.$$

- Note that if  $Q(q)$  turns out to be a **linear aggregator function**, so that  $Q(q^t) \equiv \alpha \cdot q^t = \sum_{n=1}^N \alpha_n q_{tn}$ , then the corresponding period  $t$  price level  $P^t$  is equal to  $p^t \cdot q^t / \alpha \cdot q^t$ , which is a *quality adjusted unit value price level*.

# The Assumption of Maximizing Behavior is Introduced

- Two additional assumptions are made:
  - (v)  $Q(q)$  is **once differentiable** with respect to the components of  $q$ ;
  - (vi) the observed **strictly positive quantity** vector for period  $t$ ,  $q^t \gg 0_N$ , **is a solution** to the following period  $t$  constrained maximization problem:
    - (4)  $\max_q \{Q(q) : p^t \cdot q = v^t ; q \geq 0_N\}; t = 1, \dots, T.$
- The **first order conditions** for solving (4) for period  $t$  are the following conditions:
  - (5)  $\nabla_q Q(q^t) = \lambda_t p^t ; t = 1, \dots, T;$
  - (6)  $p^t \cdot q^t = v^t ; t = 1, \dots, T.$
- This theory dates back to Konüs and Byushgens (1926), Shephard (1953) (in the context of a cost minimization framework), Samuelson and Swamy (1974) and Diewert (1976).

## Some Implications of Maximizing Behavior

- Since  $Q(q)$  is assumed to be linearly homogeneous with respect to  $q$ , **Euler's Theorem on homogeneous functions** implies that the following equations hold:

$$(7) \quad q^t \cdot \nabla_q Q(q^t) = Q(q^t) ; \quad t = 1, \dots, T.$$

- Take the inner product of both sides of equations (5) with  $q^t$  and use the resulting equations along with equations (7) to solve for the Lagrange multipliers,  $\lambda_t$ :

$$(8) \quad \lambda_t = Q(q^t) / p^t \cdot q^t \quad t = 1, \dots, T \\ = 1/P^t$$

using definitions (3).

- **Thus the Lagrange multipliers for the utility maximization problems are equal to the reciprocals of the aggregate price levels.**

## Additional Implications of Maximizing Behavior

- Thus if we assume utility maximizing behavior on the part of purchasers of the  $N$  products using the collective utility function  $Q(q)$  that satisfies the above regularity conditions, then the period  $t$  quantity aggregate is  $Q^t \equiv Q(q^t)$  and the companion period  $t$  price level defined as  $P^t \equiv p^t \cdot q^t / Q^t$  is equal to  $1/\lambda_t$  where  $\lambda_t$  is the Lagrange multiplier for problem  $t$  in the constrained utility maximization problems (4) and where  $q^t$  and  $\lambda_t$  solve equations (5) and (6) for period  $t$ .
- Equations (8) also imply that the product of  $P^t$  and  $Q^t$  is exactly equal to observed period  $t$  expenditure  $v_t$ ; i.e., we have  
(9)  $P^t Q^t = p^t \cdot q^t = v_t$  ;  $t = 1, \dots, T$ . (the **product test for levels**).
- Substitute equations (8) into equations (5) and after a bit of rearrangement, the following *fundamental equations* are obtained:  
(10)  $p^t = P^t \nabla_q Q(q^t)$  ;  $t = 1, \dots, T$ . (Note the appearance of  **$P^t$**  here).<sup>42</sup>

## The Path Forward

- In the following section, we will assume that the aggregator function,  $Q(q)$  is a linear function and we will show how this assumption along with equations (9) for the case where  $T = 2$  and  $N = 3$  can lead to a simple well known method for quality adjustment that does not involve any econometric estimation of the parameters of the linear function.
- In subsequent sections, equations (10) will be utilized in the hedonic regression context and finally, in the final sections of the paper, the assumption that  $Q(q)$  is a linear function will be relaxed.

## A Nonstochastic Method for Quality Adjustment: A Simple Model

- Consider the special case where the number of periods  $T$  is equal to 2 and the number of products in scope for the elementary index is  $N$  equal to 3.
- Product 1 is **present in both periods**, product 2 is present in period 1 but not in period 2 (**a disappearing product**) and product 3 is not present in period 1 but is present in period 3 (**a new product**).
- We assume that purchasers of the three products behave as if they collectively maximized the following **linear aggregator function**:

$$(11) Q(q_1, q_2, q_3) \equiv \alpha_1 q_1 + \alpha_2 q_2 + \alpha_3 q_3$$

- where the  $\alpha_n$  are positive constants. Under these assumptions, equations (10) written out in scalar form become the following equations:

## The Simple 3 Product, 2 Period Model

$$(12) p_{tn} = P^t \alpha_n ; \quad n = 1,2,3; t = 1,2.$$

- Equations (12) are 6 equations in the 5 parameters  $P^1$  and  $P^2$  (which can be interpreted as *aggregate price levels* for periods 1 and 2) and  $\alpha_1$ ,  $\alpha_2$  and  $\alpha_3$ , which can be interpreted as *quality adjustment factors* for the 3 products; i.e., **each  $\alpha_n$  measures the relative usefulness of an additional unit of product n to purchasers** of the 3 products.
- However, product 3 is not observed in the marketplace during period 1 and product 2 is not observed in the marketplace in period 2 and so there are **only 4 equations in (12) to determine 5 parameters**.
- However, the  $P^t$  and the  $\alpha_n$  cannot all be identified using observable data; i.e., if  $P^1$ ,  $P^2$ ,  $\alpha_1$ ,  $\alpha_2$  and  $\alpha_3$  satisfy equations (12) and  $\lambda$  is any positive number, then  $\lambda P^1$ ,  $\lambda P^2$ ,  $\lambda^{-1}\alpha_1$ ,  $\lambda^{-1}\alpha_2$  and  $\lambda^{-1}\alpha_3$  will also satisfy equations (12).
- Thus it is necessary to place a normalization (like  $P^1 = 1$  or  $\alpha_1 = 1$ ) on the 5 parameters which appear in equations (12) in order to obtain a unique solution.

## The Simple 3 Product, 2 Period Model (cont)

- In the index number context, it is natural to set the price level for period 1 equal to unity and so we impose the following **normalization** on the 5 unknown parameters which appear in equations (12):

$$(13) P^1 = 1.$$

- The 4 equations in (12) which involve observed prices and the single equation (13) are **5 equations in 5 unknowns**. The **unique solution** to these equations is:

$$(14) P^1 = 1; P^2 = p_{21}/p_{11}; \alpha_1 = p_{11}; \alpha_2 = p_{12}; \alpha_3 = p_{23}/(p_{21}/p_{11}) = p_{23}/P^2.$$

- Note that the resulting **price index**,  $P^2/P^1$ , is equal to  $p_{21}/p_{11}$ , **the price ratio for the commodity that is present in both periods**.
- Thus the price index for this very simple model turns out to be a **maximum overlap price index**.

## Reservation Prices for the Missing Prices

- Once the  $P^t$  and  $\alpha_n$  have been determined, equations (12) for the missing products can be used to define the following *imputed prices*  $p_{tn}^*$  for commodity 3 in period 1 and product 2 in period 2:

$$(15) p_{13}^* \equiv P^1 \alpha_3 = p_{23}/(P^2/P^1) ; p_{22}^* \equiv P^2 \alpha_2 = (p_{21}/p_{11})p_{12} = (P^2/P^1)p_{12}.$$

- These imputed prices can be interpreted as Hicksian (1940; 12) *reservation prices*; i.e., they are the lowest possible prices that are just high enough to deter purchasers from purchasing the products during periods if the unavailable products hypothetically became available.
- Note that  $p_{13}^* = p_{23}/(P^2/P^1)$  is an *inflation adjusted carry backward price*; i.e., the observed price for product 3 in period 2,  $p_{23}$ , is divided by the maximum overlap price index  $P^2/P^1$  in order to obtain a “reasonable” valuation for a unit of product 3 in period 1.

## Reservation Prices for the Missing Prices (cont)

- Similarly,  $p_{22}^* = (P^2/P^1)p_{12}$  is an ***inflation adjusted carry forward price*** for product 2 in period 2; i.e., the observed price for product 2 in period 1,  $p_{12}$ , is multiplied by the maximum overlap price index  $P^2/P^1$  in order to obtain a “reasonable” valuation for a unit of product 2 in period 2.
- The use of carry forward and backward prices to estimate missing prices is widespread in statistical agencies. For additional materials on this method for estimating missing prices, see Triplett (2004), de Haan and Krsinich (2012) and Diewert, Fox and Schreyer (2017).
- The simple model explained above provides a ***consumer theory justification*** for the use of these imputed prices.

## Two Methods for Computing Price and Quantity Levels

- Note that the above algebra can be implemented without a knowledge of quantities sold or purchased.
- Assuming that quantity information is available, we now consider how companion quantity levels,  $Q^1$  and  $Q^2$ , for the price levels,  $P^1 = 1$  and  $P^2$ , can be determined.
- Note that  $q_{13} = 0$  and  $q_{22} = 0$  since consumers cannot purchase products that are not available.
- Use the imputed prices defined by (15) to obtain complete price vectors for each period; i.e., define the period 1 complete price vector by  $p^1 \equiv [p_{11}, p_{12}, p_{13}^*]$  and the complete period 2 price vector by  $p^2 \equiv [p_{21}, p_{22}^*, p_{23}]$ .
- The corresponding complete quantity vectors are by  $q^1 \equiv [q_{11}, q_{12}, 0]$  and  $q^2 \equiv [q_{21}, 0, q_{23}]$ .

## Two Methods for Computing Price and Quantity Levels (cont)

- The period  $t$  aggregate quantity level  $Q^t$  can be calculated **directly** using only information on  $q^t$  and the vector of quality adjustment factors,  $\alpha \equiv [\alpha_1, \alpha_2, \alpha_3]$ , or **indirectly** by deflating period  $t$  expenditure  $v_t \equiv p^t \cdot q^t$  by the estimated period  $t$  price level,  $P^t$ .
- Thus we have the following **two possible methods for constructing the  $Q^t$** :  
(16)  $Q^t \equiv \alpha \cdot q^t$  ; or  $Q^t \equiv p^t \cdot q^t / P^t$  ;  $t = 1, 2$ .
- However, using the complete price vectors  $p^t$  with imputed prices filling in for the missing prices, equations (12) hold exactly and thus it is straightforward to show that  $Q^t = \alpha \cdot q^t = p^t \cdot q^t / P^t$  for  $t = 1, 2$ .
- Thus it does not matter whether we use the direct or indirect method for calculating the quantity levels; **both methods give the same answer in this simple model.**

## A More Complicated Model

- A problem with the above simple model is that there is only one product that is present in both periods. We need to generalize the simple model to allow for multiple overlapping products.
- In order to generalize the very simple model for dealing with new and disappearing products that was presented in the previous section, we develop another application of the fundamental equations (10).
- Define the aggregator function  $Q(q)$  as follows:  
(17)  $Q_{\text{KBF}}(q^*) \equiv [q^* \cdot A q^*]^{1/2}$
- where  $q^*$  is defined as the  $N$  dimensional quantity vector  $[q_1^*, \dots, q_N^*]$  and  $A \equiv [a_{ij}]$  is an  $N$  by  $N$  symmetric matrix of parameters which satisfies certain regularity conditions.
- Suppose further that that the observed price and quantity vectors for periods 1 and 2 are the positive price and quantity vectors,  $p^{t*} \equiv [p_{t1}^*, \dots, p_{tN}^*]$  and  $q^{t*} \equiv [q_{t1}^*, \dots, q_{tN}^*]$  for  $t = 1, 2$ .
- Why the stars? You will see why in due course!

## The KBF Model

- The model that we are about to develop is due to **Konüs and Byushgens** (1926) who showed the relationship of the KB functional form defined by (17) to the **Fisher** (1922) ideal index.
- We assume that  $q^{t*}$  solves  $\max_q \{Q(q) : p^{t*} \cdot q = v^{t*} ; q \geq 0_N\}$  for  $t = 1, 2$  where  $v^{t*} \equiv p^{t*} \cdot q^{t*}$  is observed expenditure on the  $N$  products for periods  $t = 1, 2$ .
- The inverse demand functions (10) that correspond to this particular aggregator function are the following ones:  
(18)  $p^{t*} = P^{t*} \nabla_q Q_{\text{KBF}}(q^{t*}) = P^t [q^{t*} \cdot Aq^{t*}]^{-1/2} Aq^{t*} ; \quad t = 1, 2.$
- Using the framework described in section 2 above, the period  $t$  aggregate quantity level for the present model is  $Q^{t*} \equiv [q^{t*} \cdot Aq^{t*}]^{1/2}$  and the corresponding period  $t$  price level is  $P^{t*} \equiv p^{t*} \cdot q^{t*} / Q^{t*}$  for  $t = 1, 2$ .
- In the following slide, we define the **Fisher** (1922) *ideal quantity index*:

## The KBF Model (cont)

$$(19) Q_F(p^{1*}, p^{2*}, q^{1*}, q^{2*}) \equiv [p^{1*} \cdot q^{2*} p^{2*} \cdot q^{2*} / p^{1*} \cdot q^{1*} p^{2*} \cdot q^{1*}]^{1/2}.$$

- Use equations (18) to eliminate  $p^{1*}$  and  $p^{2*}$  from the right hand side of (19). We find that

$$(20) (p^{1*} \cdot q^{2*} p^{2*} \cdot q^{2*}) / (p^{1*} \cdot q^{1*} p^{2*} \cdot q^{1*}) = q^{2*} \cdot A q^{2*} / q^{1*} \cdot A q^{1*}.$$

- Take positive square roots on both sides of (20).
- Using definitions (17) and (19), the resulting equation is:

$$(21) Q_{KBF}(q^{2*}) / Q_{KBF}(q^{1*}) = Q_F(p^{1*}, p^{2*}, q^{1*}, q^{2*}).$$

- Thus  $Q^{2*} / Q^{1*} = Q_{KBF}(q^{2*}) / Q_{KBF}(q^{1*})$  is equal to the Fisher quantity index  $Q_F(p^{1*}, p^{2*}, q^{1*}, q^{2*})$ , which can be calculated using observable price and quantity data for the two periods.
- We know from section 2 that

$$(22) P^{t*} Q^{t*} = p^{t*} \cdot q^{t*} ; t = 1, 2.$$

## The KBF Model (cont)

- Now make the normalization  $P^{1*} = 1$ . Using this normalization and equations (21) and (22), **the aggregate price and quantity levels for the two periods can be defined in terms of observable data as follows:**

$$(23) \quad P^{1*} \equiv 1; \quad Q^{1*} \equiv p^{1*} \cdot q^{1*}; \quad Q^{2*} \equiv Q^{1*} Q_F(p^{1*}, p^{2*}, q^{1*}, q^{2*}); \\ P^{2*} \equiv p^{1*} \cdot q^{1*} / Q^{2*}.$$

- The above results can be combined with the 3 product model that was described in the previous section: **relabel the above aggregate data as a composite product 1** so that the new product 1 that corresponds to the first product in section 3 has prices and quantities defined as  $p_{t1} \equiv P^{t*}$  and  $q_{t1} \equiv Q^{t*}$  for  $t = 1, 2$ .
- Products 2 and 3 are a disappearing product and a new product respectively as in section 3 above. The aggregate price levels for the two periods (which use all  $N+2$  products) are  $P^1$  and  $P^2$  and the new  $\alpha_n$  parameters are defined by the following counterparts to equations (14) above:

## A More General New and Disappearing Product Model

$$(24) P^1 = 1; P^2 = P^{2^*}/P^{1^*} = P_F(p^{1^*}, p^{2^*}, q^{1^*}, q^{2^*}); \alpha_1 = 1; \alpha_2 = p_{12};$$

$$\alpha_3 = p_{23}/(P^{2^*}/P^{1^*})$$

- where  $P^{2^*}/P^{1^*} \equiv [v^{2^*}/v^{1^*}]/[Q^{2^*}/Q^{1^*}] \equiv P_F(p^{1^*}, p^{2^*}, q^{1^*}, q^{2^*})$  is the Fisher (1922) ideal price index that compares the prices of the  $N$  products that are present in both periods,  $p^{1^*}$ ,  $p^{2^*}$ , for the two periods under consideration.
  - The imputed prices for the missing products,  $p_{13}^*$  and  $p_{22}^*$ , are obtained by using equations (15) for our present model:
- $$(25) p_{13}^* \equiv p_{23}/P_F(p^{1^*}, p^{2^*}, q^{1^*}, q^{2^*}); p_{22}^* \equiv P_F(p^{1^*}, p^{2^*}, q^{1^*}, q^{2^*})p_{12}.$$
- Comparing (24) and (25) with the corresponding equations (14) and (15) for the 3 product model, it can be seen that the price ratio for product 1 that was present in both periods,  $p_{21}/p_{11}$ , is replaced by the Fisher index  $P_F(p^{1^*}, p^{2^*}, q^{1^*}, q^{2^*})$  which is now defined over the set of products that are present in both periods.

## A More General New and Disappearing Product Model (cont)

- The type of **inflation adjusted carry backward price**  $p_{13}^*$  and the **inflation adjusted carry forward price**  $p_{22}^*$  defined by (25) are widely used by statistical agencies to estimate missing prices but usually using Laspeyres or Paasche indexes in place of the Fisher price index.
- The aggregator function that is consistent with the new model with N continuing products, one disappearing product and one new product is defined as follows:

$$(26) Q(q_1^*, \dots, q_N^*, q_2, q_3) \equiv \alpha_1 Q_{\text{KBF}}(q^*) + \alpha_2 q_2 + \alpha_3 q_3$$

- where  $Q_{\text{KBF}}(q^*)$  is the KBF aggregator function defined by (17) and  $\alpha_1$  is set equal to 1.
- Note that **the model defined by (26) is restrictive from the economic perspective because the additive nature of definition (26) implies that the composite first commodity is perfectly substitutable (after quality adjustment) with the new and disappearing commodities (which are also perfect substitutes for each other after quality adjustment).**
- However, if the products under consideration are highly substitutable for each other, the implicit assumption of perfect substitutes for missing products may be acceptable.

## Time Product Dummy Regressions: The Case of No Missing Observations and Equal Weighting

- Let  $p^t \equiv [p_{t1}, \dots, p_{tN}]$  and  $q^t \equiv [q_{t1}, \dots, q_{tN}]$  denote the price and quantity vectors for time periods  $t = 1, \dots, T$ .
- Initially, we assume that there are **no missing prices** or quantities so that all NT prices and quantities are positive.
- We assume that the quantity aggregator function  $Q(q)$  is the following **linear function**:

$$(27) Q(q) = Q(q_1, q_2, \dots, q_N) \equiv \sum_{n=1}^N \alpha_n p_n = \alpha \cdot q$$

- where the  $\alpha_n$  are positive parameters, which can be interpreted as **quality adjustment factors**.
- Under the assumption of maximizing behavior on the part of purchasers of the  $N$  commodities, assumption (27) applied to equations (10) imply that the following NT equations should hold exactly:

## Time Product Dummy Regressions (cont)

$$(28) p_{tn} = \pi_t \alpha_n ; n = 1, \dots, N; t = 1, \dots, T$$

- where we have redefined the period  $t$  price levels  $P^t$  in equations (10) as the parameters  $\pi_t$  for  $t = 1, \dots, T$ .
- Note that equations (28) form the basis for the *time dummy hedonic regression model* which is due to Court (1939). Note that these equations are a special case of the model of consumer behavior that was explained in section 2 above.
- At this point, it is necessary to point out that our consumer theory derivation of equations (28) is not accepted by all economists. Rosen (1974), Triplett (1987) and Pakes (2001) have argued for a more general approach to the derivation of hedonic regression models that is based on **supply conditions** as well as on **demand conditions**. **The present approach is obviously based on consumer demands and preferences only.**

## Time Product Dummy Regressions (cont)

- Empirically, equations (28) are unlikely to hold exactly.
- Thus we assume that the exact model defined by (28) holds only to some degree of approximation and so error terms,  $e_{tn}$ , are added to the right hand sides of equations (28).
- Here are the **two key questions** that we need to address:
  - (i) How exactly are the error terms to be introduced into the exact equations (28)?
  - (ii) Should we weight equations (28) according to their economic importance and if so, what weights should be used?
- Our approach to answering both questions will be a pragmatic one. We will experiment with different ways of introducing error terms and weights into equations (28) and reject specifications which give rise to indexes which have awkward axiomatic or economic properties.

## Time Product Dummy Regressions (cont)

- We will postpone the weighting problem for a while and look at different ways of introducing the error terms into equations (28).
- Our approach will not be very rigorous from an econometric point of view; we will simply generate different indexes as solutions to various least squares minimization problems.
- Our first approach is to simply add error terms,  $e_{tn}$ , to the right hand sides of equations (28). The unknown parameters,  $\pi \equiv [\pi_1, \dots, \pi_T]$  and  $\alpha \equiv [\alpha_1, \dots, \alpha_N]$ , will be estimated as solutions to the following (nonlinear) least squares minimization problem:

$$(29) \min_{\alpha, \pi} \sum_{n=1}^N \sum_{t=1}^T [p_{tn} - \pi_t \alpha_n]^2 .$$

- Throughout the paper, we will obtain estimators for the aggregate price levels  $\pi_t$  and the quality adjustment parameters  $\alpha_n$  as solutions to least squares minimization problems like those defined by (29) or as solutions to weighted least squares minimization problems that will be considered in subsequent sections.

# Time Product Dummy Regressions: Approach 1

- The first order necessary (and sufficient) conditions for  $\pi \equiv [\pi_1, \dots, \pi_T]$  and  $\alpha \equiv [\alpha_1, \dots, \alpha_N]$  to solve the minimization problem defined by (29) are equivalent to the following  $N + T$  equations:

$$(30) \quad \alpha_n = \sum_{t=1}^T \pi_t p_{tn} / \sum_{t=1}^T \pi_t^2; \quad n = 1, \dots, N$$

$$= \sum_{t=1}^T \pi_t^2 (p_{tn} / \pi_t) / \sum_{t=1}^T \pi_t^2 ;$$

$$(31) \quad \pi_t = \sum_{n=1}^N \alpha_n p_{tn} / \sum_{n=1}^N \alpha_n^2; \quad t = 1, \dots, T$$

$$= \sum_{n=1}^N \alpha_n^2 (p_{tn} / \alpha_n) / \sum_{n=1}^N \alpha_n^2.$$

- Solutions to the two sets of equations can readily be obtained by iterating between the two sets of equations.
- If  $\pi^* \equiv [\pi_1^*, \dots, \pi_T^*]$  and  $\alpha^* \equiv [\alpha_1^*, \dots, \alpha_N^*]$  is a solution to (30) and (31), then  $\lambda \pi^*$  and  $\lambda^{-1} \alpha^*$  is also a solution for any  $\lambda > 0$ . Thus **to obtain a unique solution we impose the normalization  $\pi_1^* = 1$ .**
- Then  $1, \pi_2^*, \dots, \pi_T^*$  is the sequence of price levels that is generated by the least squares minimization problem defined by (29).

## Time Product Dummy Regressions: Approach 1 (cont)

- If quantity data are available, then using the general methodology that was outlined in section 2, aggregate quantity levels for the  $t$  periods can be obtained as  $Q^{t*} \equiv \alpha^* \cdot q^t = \sum_{n=1}^N \alpha_n^* q_{tn}$  for  $t = 1, \dots, T$ .
- Estimated aggregate price levels can be obtained **directly** from the solution to (29); i.e., set  $P^{t*} = \pi_t^*$  for  $t = 1, \dots, T$ .
- Alternative price levels can be **indirectly** obtained as  $P^{t**} \equiv p^t \cdot q^t / Q^{t*} = p^t \cdot q^t / \alpha^* \cdot q^t$  for  $t = 1, \dots, T$ .
- If the optimized objective function in (29) is 0 (so that all errors  $e_{tn}^* \equiv p_{tn} - \pi_t^* \alpha_n^*$  equal 0 for  $t = 1, \dots, T$  and  $n = 1, \dots, N$ ), then  $P^{t*}$  will equal  $P^{t**}$  for all  $t$ .
- **However, usually nonzero errors will occur and so a choice between the two sets of estimators must be made.**

## Time Product Dummy Regressions: Approach 1 Rejected

- From (30), it can be seen that  $\alpha_n^*$ , the quality adjustment parameter for product n, is a weighted average of the T inflation adjusted prices for product n, the  $p_{tn}/\pi_t^*$ , where the weight for  $p_{tn}/\pi_t^*$  is  $\pi_t^{*2} / \sum_{\tau=1}^T \pi_{\tau}^{*2}$ . This means that the weight for  $p_{tn}/\pi_t^*$  will be very high for periods t where general inflation is high, which seems rather arbitrary.
- In addition to having unattractive weighting properties, the estimates generated by solving the least squares minimization problem (29) suffer from a fatal flaw: *the estimates are not invariant to changes in the units of measurement.*
- In order to remedy this defect, we turn to an alternative error specification.

## Time Product Dummy Regressions: Approach 2

- Instead of adding approximation errors to the exact equations (28), we could append multiplicative approximation errors. Thus the exact equations become  $p_{tn} = \pi_t \alpha_n e_{tn}$  for  $n = 1, \dots, N$  and  $t = 1, \dots, T$ . Upon taking logarithms of both sides of these equations, we obtain the following system of estimating equations:

$$(32) \ln p_{tn} = \ln \pi_t + \ln \alpha_n + \ln e_{tn} ; \quad n = 1, \dots, N; t = 1, \dots, T \\ = \rho_t + \beta_n + \varepsilon_{tn}$$

- where  $\rho_t \equiv \ln \pi_t$  for  $t = 1, \dots, T$  and  $\beta_n \equiv \ln \alpha_n$  for  $n = 1, \dots, N$ .
- The model defined by (32) is the basic *Time Product Dummy regression model with no missing observations*.
- Now choose the  $\rho_t$  and  $\beta_n$  to minimize the sum of squared residuals,  $\sum_{n=1}^N \sum_{t=1}^T \varepsilon_{tn}^2$ . Thus let  $\rho \equiv [\rho_1, \dots, \rho_T]$  and  $\beta \equiv [\beta_1, \dots, \beta_N]$  be a solution to the following least squares minimization problem:

$$(33) \min_{\rho, \beta} \sum_{n=1}^N \sum_{t=1}^T [\ln p_{tn} - \rho_t - \beta_n]^2 .$$

## Time Product Dummy Regressions: Approach 2 (cont)

- The first order necessary conditions for  $\rho_1, \dots, \rho_T$  and  $\beta_1, \dots, \beta_N$  to solve (33) are the following  $T + N$  equations:

$$(34) \quad N\rho_t + \sum_{n=1}^N \beta_n = \sum_{n=1}^N \ln p_{tn} ; \quad t = 1, \dots, T;$$

$$(35) \quad \sum_{t=1}^T \rho_t + T\beta_n = \sum_{t=1}^T \ln p_{tn} ; \quad n = 1, \dots, N.$$

- Replace the  $\rho_t$  and  $\beta_n$  in equations (34) and (35) by  $\ln \pi_t$  and  $\ln \alpha_n$  respectively for  $t = 1, \dots, T$  and  $n = 1, \dots, N$ . After some rearrangement, the resulting equations become:

$$(36) \quad \pi_t = \prod_{n=1}^N (p_{tn}/\alpha_n)^{1/N} ; \quad t = 1, \dots, T;$$

$$(37) \quad \alpha_n = \prod_{t=1}^T (p_{tn}/\pi_t)^{1/T} ; \quad n = 1, \dots, N.$$

- Thus the period  $t$  aggregate price level,  $\pi_t$ , is equal to the geometric average of the  $N$  quality adjusted prices for period  $t$ ,  $p_{t1}/\alpha_1, \dots, p_{tN}/\alpha_N$ , while the quality adjustment factor for product  $n$ ,  $\alpha_n$ , is equal to the geometric average of the  $T$  inflation adjusted prices for product  $n$ ,  $p_{1n}/\pi_1, \dots, p_{Tn}/\pi_T$ .
- These estimators look very reasonable (if quantity weights are not available).

## Time Product Dummy Regressions: Approach 2 (cont)

- If  $\pi^* \equiv [\pi_1^*, \dots, \pi_T^*]$  and  $\alpha^* \equiv [\alpha_1^*, \dots, \alpha_N^*]$  is a solution to (36) and (37), then  $\lambda\pi^*$  and  $\lambda^{-1}\alpha^*$  is also a solution for any  $\lambda > 0$ . Thus to obtain a unique solution we impose the normalization  $\pi_1^* = 1$  (which corresponds to  $\rho_1 = 0$ ).
- Then  $1, \pi_2^*, \dots, \pi_T^*$  is the **sequence of price levels** that is generated by the least squares minimization problem defined by (33).
- Once we have the unique solution  $1, \pi_2^*, \dots, \pi_T^*$  for the T price levels that are generated by the (33), the **price index** between period t relative to period s can be defined as  $\pi_t^*/\pi_s^*$ .
- Using equations (36) for  $\pi_t^*$  and  $\pi_s^*$ , we have the following expression for the **price index**:
 
$$(38) \quad \pi_t^*/\pi_s^* = \prod_{n=1}^N (p_{tn}/\alpha_n^*)^{1/N} / \prod_{n=1}^N (p_{sn}/\alpha_n^*)^{1/N}$$

$$= \prod_{n=1}^N (p_{tn}/p_{sn})^{1/N}.$$
- This is simply the Jevons index for period t relative to period s.

## Time Product Dummy Regressions: Approach 2 (conc)

- Thus if there are no missing observations, the Time Product Dummy price indexes between any two periods in the window of T period under consideration is equal to the *Jevons index* between the two periods (the simple geometric mean of the price ratios,  $p_{tn}/p_{sn}$ ).
- This is a somewhat **disappointing result** since an equally weighted average of the price ratios is not necessarily a representative average of the prices; i.e., unimportant products to purchasers (in the sense that they spend very little on these products) are given the same weight in the Jevons measure of inflation between the two periods as is given to high expenditure products.
- This result indicates the importance of weighting.
- In the next section, I look at Approach 2 when there are missing observations.

## 6. Time Product Dummy Regressions: The Case of Missing Observations with no Weighting by Economic Importance

- As in the previous section, there are  $N$  products and  $T$  time periods but not all products are purchased (or sold) in all time periods. For each period  $t$ , **define the set of products  $n$  that are present in period  $t$  as  $S(t) \equiv \{n: p_{tn} > 0\}$  for  $t = 1, 2, \dots, T$ . It is assumed that these sets are not empty; i.e., at least one product is purchased in each period.**
- For each product  $n$ , **define the set of periods  $t$  where product  $n$  is present as  $S^*(n) \equiv \{t: p_{tn} > 0\}$ . Again, assume that these sets are not empty; i.e., each product is sold in at least one time period.**
- Define the integers  $N(t)$  and  $T(n)$  as follows:  
(40)  $N(t) \equiv \sum_{n \in S(t)} 1; \quad t = 1, \dots, T;$   
(41)  $T(n) \equiv \sum_{t \in S^*(n)} 1; \quad n = 1, \dots, N.$
- If all  $N$  products are present in period  $t$ , then  $N(t) = N$ ; if product  $n$  is present in all  $T$  periods, then  $T(n) = T$ .

## 6. Time Product Dummy Regressions with Missing Observations

- Using the notation that was defined in the previous section, the counterpart to (33) when there are missing products is the following least squares minimization problem:

$$(42) \min_{\rho, \beta} \sum_{t=1}^T \sum_{n \in S(t)} [\ln p_{tn} - \rho_t - \beta_n]^2$$

$$= \min_{\rho, \alpha} \sum_{n=1}^N \sum_{t \in S^*(n)} [\ln p_{tn} - \rho_t - \beta_n]^2.$$

- Note that there are two equivalent ways of writing the least squares minimization problem.
- The first order necessary conditions for  $\rho_1, \dots, \rho_T$  and  $\beta_1, \dots, \beta_N$  to solve (42) are the following counterparts to (34) and (35):

$$(43) \sum_{n \in S(t)} [\rho_t + \beta_n] = \sum_{n \in S(t)} \ln p_{tn} ; \quad t = 1, \dots, T;$$

$$(44) \sum_{t \in S^*(n)} [\rho_t + \beta_n] = \sum_{t \in S^*(n)} \ln p_{tn} ; \quad n = 1, \dots, N.$$

- Let  $\rho_t \equiv \ln \pi_t$  for  $t = 1, \dots, T$  and let  $\beta_n \equiv \ln \alpha_n$  for  $n = 1, \dots, N$ .
- Substitute these definitions into equations (43) and (44). After some rearrangement and using definitions (40) and (41), equations (43) and (44) become the following ones:

## 6. Time Product Dummy Regressions with Missing Observations

$$(45) \pi_t = \prod_{n \in S(t)} [p_{tn}/\alpha_n]^{1/N(t)} ; \quad t = 1, \dots, T;$$

$$(46) \alpha_n = \prod_{t \in S^*(n)} [p_{tn}/\pi_t]^{1/T(n)} ; \quad n = 1, \dots, N.$$

- To obtain a unique solution we impose the normalization  $\pi_1^* = 1$  (which corresponds to  $\rho_1 = 0$ ).
- Then  $1, \pi_2^*, \dots, \pi_T^*$  is the sequence of (normalized) price levels that is generated by the least squares minimization problem defined by (42).
- In this case,  $\pi_t^* = \prod_{n \in S(t)} [p_{tn}/\alpha_n^*]^{1/N(t)}$  is the equally weighted geometric mean of all of the quality adjusted prices for the products that are available in period  $t$  for  $t = 2, 3, \dots, T$ .
- We have the following expressions for  $\pi_t^*/\pi_r^*$  and  $\alpha_n^*/\alpha_m^*$ :
 
$$(47) \pi_t^*/\pi_r^* = \prod_{n \in S(t)} [p_{tn}/\alpha_n^*]^{1/N(t)} / \prod_{n \in S(r)} [p_{rn}/\alpha_n^*]^{1/N(r)} ; \quad 1 \leq t, r \leq T;$$

$$(48) \alpha_n^*/\alpha_m^* = \prod_{t \in S^*(n)} [p_{tn}/\pi_t^*]^{1/T(n)} / \prod_{t \in S^*(m)} [p_{tm}/\pi_t^*]^{1/T(m)} ; 1 \leq n, m \leq N.$$
- **Now the  $\alpha_n^*$  enter into the indexes  $\pi_t^*/\pi_r^*$  defined by (47).**

## 6. Time Product Dummy Regressions with Missing Observations

- If the set of available products is the same in periods  $r$  and  $t$ , then the quality adjustment factors do cancel and the price index for period  $t$  relative to period  $r$  is  $\pi_t^* / \pi_r^* = \prod_{n \in S(t)} [p_{tn}/p_{rn}]^{1/N(t)}$ , which is the Jevons index between periods  $r$  and  $t$ .
- Again, while this index is an excellent one if quantity information is not available, it is not satisfactory when quantity information is available due to its equal weighting of economically important and unimportant price ratios.
- There is another unsatisfactory property of the estimated price levels that are generated by solving the time product dummy hedonic model that is defined by (42): a product that is available only in one period out of the  $T$  periods has no influence on the aggregate price levels  $\pi_t^*$ .
- We turn to hedonic models which involve weighting.

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 1

- In order to take economic importance into account, for the case of 2 time periods, replace (33) by the following *weighted least squares minimization problem*:

$$(51) \min_{\rho, \beta} \sum_{n=1}^N q_{1n} [\ln p_{1n} - \beta_n]^2 + \sum_{n=1}^N q_{2n} [\ln p_{2n} - \rho_2 - \beta_n]^2$$

- where we have set  $\rho_1 = 0$ .
  - But this weighted model generates the following solution for  $\rho_2$
- $$(54) \rho_2^* \equiv \sum_{n=1}^N q_{1n} q_{2n} (q_{1n} + q_{2n})^{-1} \ln(p_{2n}/p_{1n}) / \sum_{i=1}^N q_{1i} q_{2i} (q_{1i} + q_{2i})^{-1}.$$
- The resulting  $\pi_2^* = \exp[\rho_2^*]$  is not invariant to changes in the units of measurement; **this index is not satisfactory!**
  - Now replace the quantity weights in (51) with value weights which leads to Approach 2:

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 2

- Since values are invariant to changes in the units of measurement, the lack of invariance problem could be solved if we **replace the quantity weights in (51) with expenditure or sales weights**. This leads to the following weighted least squares minimization problem where the weights  $v_{tn}$  are defined as  $p_{tn}q_{tn}$  for  $t = 1,2$  and  $n = 1,\dots,N$ :

$$(58) \min_{\rho, \beta} \sum_{n=1}^N v_{1n} [\ln p_{1n} - \beta_n]^2 + \sum_{n=1}^N v_{2n} [\ln p_{2n} - \rho_2 - \beta_n]^2.$$

- It can be seen that problem (58) has exactly the same mathematical form as problem (51) except that  $v_{tn}$  has replaced  $q_{tn}$  and so the solutions (54) and (55) will be valid in the present context if  $v_{tn}$  replaces  $q_{tn}$  in these formulae. Thus the solution to (58) is:

$$(59) \rho_2^* \equiv \sum_{n=1}^N v_{1n} v_{2n} (v_{1n} + v_{2n})^{-1} \ln(p_{2n}/p_{1n}) / \sum_{i=1}^N v_{1i} v_{2i} (v_{1i} + v_{2i})^{-1};$$

$$(60) \beta_n^* \equiv v_{1n} (v_{1n} + v_{2n})^{-1} \ln(p_{1n}) + v_{2n} (v_{1n} + v_{2n})^{-1} \ln(p_{2n}/\pi_2^*);$$

$n = 1,\dots,N$

- where  $\pi_2^* \equiv \exp[\rho_2^*]$ .

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 2

- The resulting price index,  $\pi_2^*/\pi_1^* = \pi_2^* = \exp[\rho_2^*]$  is indeed invariant to changes in the units of measurement.
- However, if we regard  $\pi_2^*$  as a function of the price and quantity vectors for the two periods, say  $P(p^1, p^2, q^1, q^2)$ , then another problem emerges for the price index defined by the solution to (58):  $P(p^1, p^2, q^1, q^2)$  is not homogeneous of degree 0 in the components of  $q^1$  or in the components of  $q^2$ . These properties are important because it is desirable that the companion implicit quantity index defined as  $Q(p^1, p^2, q^1, q^2) \equiv [p^2 \cdot q^2 / p^1 \cdot q^1] / P(p^1, p^2, q^1, q^2)$  be homogeneous of degree 1 in the components of  $q^2$  and homogeneous of degree minus 1 in the components of  $q^1$ .
- We also want  $P(p^1, p^2, q^1, q^2)$  to be homogeneous of degree 1 in the components of  $p^2$  and homogeneous of degree minus 1 in the components of  $p^1$  and these properties are also not satisfied. Thus we conclude that the solution to the weighted least squares problem defined by (58) does not generate a satisfactory price index formula.
- **Weighting Approach 2 also fails.**

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 3

- The above deficiencies with Approach 2 can be remedied if the *expenditure amounts*  $v_{tn}$  in (58) are replaced by *expenditure shares*,  $s_{tn}$ , where  $v_t \equiv \sum_{n=1}^N v_{tn}$  for  $t = 1,2$  and  $s_{tn} \equiv v_{tn}/v_t$  for  $t = 1,2$  and  $n = 1,\dots,N$ .
- This replacement leads to the following weighted least squares minimization problem:

$$(61) \min_{\rho, \beta} \sum_{n=1}^N s_{1n} [\ln p_{1n} - \beta_n]^2 + \sum_{n=1}^N s_{2n} [\ln p_{2n} - \rho_2 - \beta_n]^2.$$

- The solution to (61) is given by:

$$(62) \rho_2^* \equiv \sum_{n=1}^N s_{1n} s_{2n} (s_{1n} + s_{2n})^{-1} \ln(p_{2n}/p_{1n}) / \sum_{i=1}^N s_{1i} s_{2i} (s_{1i} + s_{2i})^{-1};$$

$$(63) \beta_n^* \equiv s_{1n} (s_{1n} + s_{2n})^{-1} \ln(p_{1n}) + s_{2n} (s_{1n} + s_{2n})^{-1} \ln(p_{2n}/\pi_2^*); \quad n = 1,\dots,N.$$

- where  $\pi_2^* \equiv \exp[\rho_2^*]$ .
- This index has satisfactory properties. It will tend to give a somewhat lower index value  $\pi_2^*$  than the superlative Törnqvist-Theil index which it approximates to the second order.

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 3 Concluded

- From equation (38), the unweighted bilateral time product dummy regression model generated the Jevons index as the solution to the unweighted least squares minimization problem that is a counterpart to the weighted problem defined by (61) above.
- Thus appropriate weighting of the squared errors has changed the solution index dramatically: the index defined by (64) weights products by their economic importance and has good test properties whereas the Jevons index can generate very problematic results due to its lack of weighting according to economic importance.
- Note that both models have the same underlying structure; i.e., they assume that  $p_{tn}$  is approximately equal to  $\pi_t \alpha_n$  for  $t = 1, 2$  and  $n = 1, \dots, N$ . *Thus weighting by economic importance has converted a least squares minimization problem that generates a rather poor price index into a problem that generates a rather good index.*

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 4

- There is **one more weighting scheme** that generates an even better index in the bilateral context where we are running a time product dummy hedonic regression using the price and quantity data for only two periods. Consider the following weighted least squares minimization problem:

$$(65) \min_{\rho, \beta} \sum_{n=1}^N (1/2)(s_{1n}+s_{2n})[\ln p_{1n} - \beta_n]^2 \\ + \sum_{n=1}^N (1/2)(s_{1n}+s_{2n})[\ln p_{2n} - \rho_2 - \beta_n]^2.$$

- The solution to (65) simplifies to the following solution:

$$(66) \rho_2^* \equiv \sum_{n=1}^N (1/2)(s_{1n}+s_{2n})\ln(p_{2n}/p_{1n});$$

$$(67) \beta_n^* \equiv (1/2)\ln(p_{1n}) + (1/2)\ln(p_{2n}/\pi_2^*); \quad n = 1, \dots, N$$

- where  $\pi_2^* \equiv \exp[\rho_2^*]$  and  $\pi_1^* \equiv \exp[\rho_1^*] = \exp[0] = 1$  since we have set  $\rho_1^* = 0$ . **Thus the bilateral index number formula which emerges from the solution to (65) is  $\pi_2^*/\pi_1^* = P_T(p^1, p^2, q^1, q^2)$ , which is the Törnqvist (1936) Theil (1967; 137-138) index.**

## 7. Weighted Time Product Dummy Regressions: The Bilateral Case: Approach 4

- Thus the use of the weights in (65) has generated an even better bilateral index number formula than the formula which resulted from the use of the weights in (61).
- Note that all of the models in this section have the same underlying structure; i.e., they assume that  $p_{tn}$  is approximately equal to  $\pi_t \alpha_n$  for  $t = 1, 2$  and  $n = 1, \dots, N$ . But the indexes that result from alternative forms of weighting can be very different.
- This result reinforces the case for using appropriately weighted versions of the basic time product dummy hedonic regression model.
- Similar comments apply to more general hedonic regressions that use information on characteristics: in general, it is preferable to use share weights in these regressions.

## 8. Weighted Time Product Dummy Regressions: The Bilateral Case with Missing Observations

- Assume that there are products that are missing in period 2 that were present in period 1 and some new products that appear in period 2.
- The **new weighted least squares minimization problem** is:

$$(68) \min_{\rho, \beta} \sum_{n \in S(1)} s_{1n} [\ln p_{1n} - \beta_n]^2 + \sum_{n \in S(2)} s_{2n} [\ln p_{2n} - \rho_2 - \beta_n]^2.$$

- The **first order conditions** for  $\rho_2^*, \beta_1^*, \dots, \beta_N^*$  to solve (68) are equivalent to the following equations:

$$(69) \sum_{n \in S(2)} s_{2n} [\ln p_{2n} - \rho_2^* - \beta_n^*] = 0 ;$$

$$(70) (s_{1n} + s_{2n})\beta_n^* = s_{1n} \ln p_{1n} + s_{2n} [\ln p_{1n} - \rho_2^*]; \quad n \in S(1) \cap S(2);$$

$$(71) \quad \beta_n^* = \ln p_{1n} ; \quad n \in S(1), n \notin S(2);$$

$$(72) \beta_n^* = \ln p_{2n} - \rho_2^* ; \quad n \in S(2), n \notin S(1).$$

- Define the **intersection set of products**  $S^*$  as follows:

$$(73) S^* \equiv S(1) \cap S(2).$$

## 8. Weighted Time Product Dummy Regressions: The Bilateral Case with Missing Observations (cont)

- The solution to the FO conditions is the following one:

$$(76) \rho_2^* \equiv [\sum_{n \in S^*} s_{1n} s_{2n} (s_{1n} + s_{2n})^{-1} \ln(p_{2n}/p_{1n})] / [\sum_{i \in S^*} s_{1i} s_{2i} (s_{1i} + s_{2i})^{-1}] ;$$

$$(77) \beta_n^* \equiv s_{1n} (s_{1n} + s_{2n})^{-1} \ln(p_{1n}) + s_{2n} (s_{1n} + s_{2n})^{-1} \ln(p_{2n}/\pi_2^*) ;$$

$$n \in S^*$$

- where  $\pi_2^* \equiv \exp[\rho_2^*]$ . Define the *normalized harmonic mean share weights* for the always present products as follows as
  - $s_n^* \equiv h(s_{1n}, s_{2n}) / \sum_{i \in S^*} h(s_{1i}, s_{2i})$  for  $n \in S^*$ .
  - Denote the period  $t$  price and quantity vectors that include only matched products by  $p^{t*}$  and  $q^{t*}$  respectively for  $t = 1, 2$ . Then the *weighted time product dummy bilateral price index with missing observations*,  $P_{WTPD}(p^{1*}, p^{2*}, q^{1*}, q^{2*}) \equiv \pi_2^* / \pi_1^* = \pi_2^*$ , has the following logarithm:
- $$(78) \ln P_{WTPD}(p^{1*}, p^{2*}, q^{1*}, q^{2*}) \equiv \sum_{n \in S^*} s_n^* \ln(p_{2n}/p_{1n}).$$
- Note the similarity of  $P_{WTPD}$  to the Törnqvist-Theil index.

## Weighted Time Product Dummy Regressions: The Bilateral Case with Missing Observations (cont)

- Note that  $P_{\text{WTPD}} \equiv \pi_2^*/\pi_1^*$  depends only on the price and share information for the products *that are present in both periods*.
- Thus the bilateral price index that is generated by solving (68) is similar to the nonstochastic **maximum overlap price index** that was defined in section 4.
- The main difference is that the maximum overlap **Fisher** price index  $P_F$  that was used in section 4 is **superlative** whereas the present index,  $P_{\text{WTPD}}$ , is only **pseudo-superlative**.
- Thus  $P_{\text{WTPD}}(p^{1*}, p^{2*}, q^{1*}, q^{2*})$  approximates  $P_F(p^{1*}, p^{2*}, q^{1*}, q^{2*})$  to the second order around any point where  $p^{1*} = p^{2*}$  and  $q^{1*} = q^{2*}$ .
- However, if the products under consideration are subject to frequent price discounts, the fluctuations in prices and quantities can be huge and second order approximations may not be very accurate.
- But normally, the section 4 and section 8 approaches will give pretty similar answers.

## Weighted Time Product Dummy Regressions: The Bilateral Case with Missing Observations (cont)

- To complete the solution to (68), use equations (71) and (72) to define the  $\beta_n^*$  for the new and missing products in period 2.
- As usual, the hedonic regression model that is generated by solving (68) can be used to **impute reservation prices for missing observations**.

- Thus define  $\alpha_n^* \equiv \exp[\beta_n^*]$  for  $n = 1, \dots, N$ .

- Then the **missing prices**  $p_{tn}^*$  can be defined as follows:

$$(79) \quad p_{2n}^* \equiv \pi_2^* \alpha_n^* = \pi_2^* p_{1n} ; \quad n \in S(1), n \notin S(2);$$

$$(80) \quad p_{1n}^* \equiv \pi_1^* \alpha_n^* = p_{2n} / \pi_2^* ; \quad n \in S(2), n \notin S(1).$$

- Thus the missing prices for period 2,  $p_{2n}^*$ , are the corresponding ***inflation adjusted carry forward prices*** from period 1,  $p_{1n}$  times  $\pi_2^*$  and the missing prices for period 1,  $p_{1n}^*$ , are the corresponding ***inflation adjusted carry backward prices*** from period 2,  $p_{2n}$  deflated by  $\pi_2^*$ .

## Weighted Time Product Dummy Regressions: The Bilateral Case with Missing Observations (cont)

- **Estimated aggregate price levels** can be obtained directly from the solution to (75); i.e., set  $P^{1*} = 1$  and  $P^{2*} = \pi_2^*$ .
- The **corresponding quantity levels** are defined as  $Q^{1*} \equiv p^1 \cdot q^1$  and  $Q^{2*} \equiv p^2 \cdot q^2 / \pi_2^*$ .
- **Alternative price and quantity levels** can be obtained as  $Q^{t**} \equiv \alpha^* \cdot q^t$  and  $P^{t**} \equiv p^t \cdot q^t / Q^{t**}$  for  $t = 1, 2$ .
- If the optimized objective function in (75) is 0, so that all errors equal 0, then  $P^{t*}$  will equal  $P^{t**}$  for all  $t$ .
- If the estimated residuals are not all equal to 0, then the two estimates for the period  $t$  price level  $P^t$  will differ and, as usual, the alternative estimates for  $P^t$  will generate different estimates for the companion aggregate quantity levels.
- This algebra brings home the not well recognized fact that basically, **hedonic regression methods assume that the products under consideration are perfect substitutes** (if we take the economic approach to index number theory).

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods

- We first consider the case of no missing observations. The generalization of the two period **weighted least squares minimization problem** that was defined by (61) in section 7 to the case of  $T > 2$  periods is (105) below:

$$(105) \min_{\rho, \beta} \sum_{n=1}^N \sum_{t=1}^T s_{tn} [\ln p_{tn} - \rho_t - \beta_n]^2.$$

- The **first order necessary conditions** for  $\rho^* \equiv [\rho_1^*, \dots, \rho_T^*]$  and  $\beta^* \equiv [\beta_1^*, \dots, \beta_N^*]$  to solve (105) are the following  $T$  equations (106) and  $N$  equations (107):

$$(106) \rho_t^* = \sum_{n=1}^N s_{tn} [\ln p_{tn}^* - \beta_n^*]; \quad t = 1, \dots, T;$$

$$(107) \beta_n^* = \sum_{t=1}^T s_{tn} [\ln p_{tn}^* - \rho_t^*] / (\sum_{t=1}^T s_{tn}); \quad n = 1, \dots, N.$$

- This method is due to Rao (1995) (2004) (2005; 574). Balk (1980; 70) suggested a similar model using different share weights.
- We set  $\rho_1^* = 0$  in equations (107) and drop the first equation in (106) and use linear algebra to find a unique solution for the resulting equations; i.e., we have the usual normalization of parameters problem.

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods (cont)

- Once the solution is found, define the estimated *price levels*  $\pi_t^*$  and *quality adjustment factors*  $\alpha_n^*$  as follows:

$$(108) \pi_t^* \equiv \exp[\rho_t^*] ; t = 1, \dots, T; \alpha_n^* \equiv \exp[\beta_n^*] ; n = 1, \dots, N.$$

- Note that the resulting *price index* between periods  $t$  and  $\tau$  is equal to the following expression:

$$(109) \pi_t^*/\pi_\tau^* = \prod_{n=1}^N \exp[s_{tn} \ln(p_{tn}/\alpha_n^*)] / \prod_{n=1}^N \exp[s_{\tau n} \ln(p_{\tau n}/\alpha_n^*)] ; \\ 1 \leq t, \tau \leq T.$$

- If  $s_{tn} = s_{\tau n}$  for  $n = 1, \dots, N$ , then  $\pi_t^*/\pi_\tau^*$  will equal a **weighted geometric mean of the price ratios**  $p_{tn}/p_{\tau n}$  where the weight for  $p_{tn}/p_{\tau n}$  is the common expenditure share  $s_{tn} = s_{\tau n}$ .
- Thus  $\pi_t^*/\pi_\tau^*$  will not depend on the  $\alpha_n^*$  in this case.
- The price levels  $\pi_t^*$  defined by (108) are functions of the  $T$  price vectors,  $p^1, \dots, p^T$  and the  $t$  quantity vectors  $q^1, \dots, q^T$ .

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods (cont)

- These price level functions have some good **axiomatic properties**:
- (i) the  $\pi_t^*$  are **invariant to changes in the units of measurement**;
- (ii)  $\pi_t^*$  regarded as a function of the period  $t$  price vector  $\mathbf{p}^t$  is **linearly homogeneous in the components of  $\mathbf{p}^t$** ; i.e.,  $\pi_t^*(\lambda \mathbf{p}^t) = \lambda \pi_t^*(\mathbf{p}^t)$  for all  $\mathbf{p}^t \gg \mathbf{0}_N$  and  $\lambda > 0$ ;
- (iii)  $\pi_t^*$  regarded as a function of the period  $t$  quantity vector  $\mathbf{q}^t$  is **homogeneous of degree 0** in the components of  $\mathbf{q}^t$ ; i.e.,  $\pi_t^*(\lambda \mathbf{q}^t) = \pi_t^*(\mathbf{q}^t)$  for all  $\mathbf{q}^t \gg \mathbf{0}_N$  and  $\lambda > 0$ ;
- (iv) the  $\pi_t^*$  satisfy a version of Walsh's (1901; 389) (1921b; 540) **multi-period identity test**; i.e., if  $\mathbf{p}^t = \mathbf{p}^\tau$  and  $\mathbf{q}^t = \mathbf{q}^\tau$ , then  $\pi_t^* = \pi_\tau^*$ .
- However if the stronger version of Walsh's multi-period identity test is not satisfied. This is why I prefer price similarity linking of observations over this method for constructing index numbers.

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods (cont)

- Once the estimates for the  $\pi_t$  and  $\alpha_n$  have been computed, we have the **usual two methods for constructing period by period price and quantity levels**,  $P^t$  and  $Q^t$  for  $t = 1, \dots, T$ .
- The  $\pi_t^*$  estimates can be used to form the aggregates using equations (110) or the  $\alpha_n^*$  estimates can be used to form the aggregates using equations (111):

$$(110) P^{t*} \equiv \pi_t^* ; \quad Q^{t*} \equiv p^t \cdot q^t / \pi_t^* ; t = 1, \dots, T;$$

$$(111) Q^{t**} \equiv \alpha^* \cdot q^t ; P^{t**} \equiv p^t \cdot q^t / \alpha^* \cdot q^t ; t = 1, \dots, T.$$

- Note that (111) tells us that the underlying price index is a **quality adjusted unit value price index** of the type studied by de Haan (2004b).
- Define the error terms  $e_{tn} \equiv \ln p_{tn} - \ln \pi_t^* - \ln \alpha_n^*$  for  $t = 1, \dots, T$  and  $n = 1, \dots, N$ . If all  $e_{tn} = 0$ , then  $P^{t*}$  will equal  $P^{t**}$  and  $Q^{t*}$  will equal  $Q^{t**}$  for  $t = 1, \dots, T$ .

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods and Missing Observations

- It is straightforward to generalize the weighted least squares minimization problem (105) to the case where there are missing prices and quantities. Here is the **generalized weighted least squares minimization problem**:

$$(112) \min_{\rho, \beta} \sum_{t=1}^T \sum_{n \in S(t)} s_{tn} [\ln p_{tn} - \rho_t - \beta_n]^2 \\ = \min_{\rho, \beta} \sum_{n=1}^N \sum_{t \in S^*(n)} s_{tn} [\ln p_{tn} - \rho_t - \beta_n]^2.$$

- Note that there are two equivalent ways of writing the least squares minimization problem.
- The **first order necessary conditions** for  $\rho_1, \dots, \rho_T$  and  $\beta_1, \dots, \beta_N$  to solve (112) are the following counterparts to (106) and (107):

$$(113) \sum_{n \in S(t)} s_{tn} [\rho_t^* + \beta_n^*] = \sum_{n \in S(t)} s_{tn} \ln p_{tn} ; t = 1, \dots, T;$$

$$(114) \sum_{t \in S^*(n)} s_{tn} [\rho_t^* + \beta_n^*] = \sum_{t \in S^*(n)} s_{tn} \ln p_{tn} ; n = 1, \dots, N.$$

- We set  $\rho_1^* = 0$  in equations (114) and drop the first equation in (113) and use linear algebra to find a unique solution for the resulting equations.

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods and Missing Observations (cont)

- Define the estimated *price levels*  $\pi_t^*$  and *quality adjustment factors*  $\alpha_n^*$  by definitions (108).

$$(108) \pi_t^* \equiv \exp[\rho_t^*] ; t = 1, \dots, T; \alpha_n^* \equiv \exp[\beta_n^*] ; n = 1, \dots, N.$$

- Substitute these definitions into equations (113) and (114).
- After some rearrangement, equations (113) and (114) become the following ones:

$$(115) \pi_t^* = \exp[\sum_{n \in S(t)} s_{tn} \ln(p_{tn}/\alpha_n^*)] ; \quad t = 1, \dots, T;$$

$$(116) \alpha_n^* = \exp[\sum_{t \in S^*(n)} s_{tn} \ln(p_{tn}/\pi_t^*) / \sum_{t \in S^*(n)} s_{tn}] ; \quad n = 1, \dots, N.$$

- Once the estimates for the  $\pi_t$  and  $\alpha_n$  have been computed, **we have the usual two methods for constructing period by period price and quantity levels,  $P^t$  and  $Q^t$  for  $t = 1, \dots, T$ ; see (110) and (111) above.**
- Again, the price indexes which result from the use of this method are quality adjusted unit value indexes.

## 9. Weighted Time Product Dummy Regressions: The General Case of Many Time Periods and Missing Observations (concluded)

- If the estimated errors  $e_{tn}^* \equiv \ln p_{tn} - \rho_t^* - \beta_n^*$  that implicitly appear in the weighted least squares minimization problem (112) turn out to be small, then the underlying exact model,  $p_{tn} = \pi_t \alpha_n$  for  $t = 1, \dots, T$ ,  $n \in S(t)$ , provides a good approximation to reality and thus this weighted time product dummy hedonic regression model can be used with some confidence.
- If the fit of the model is not good, then it may be necessary to look at other models such as the **estimation of reservation prices** or the **use of product characteristics information** or the **use of price similarity linking methods**.
- In the following section, we discuss hedonic regressions when information on product characteristics is available.

## 10. The Time Dummy Hedonic Regression Model with Characteristics Information

- In this section, it is again assumed that there are  $N$  products that are available over a window of  $T$  periods.
- As in the previous sections, we again assume that the **quantity aggregator function** for the  $N$  products is the **linear function**,  $Q(q) \equiv \alpha \cdot q = \sum_{n=1}^N \alpha_n q_n$  where  $q_n$  is the quantity of product  $n$  purchased or sold in the period under consideration and  $\alpha_n$  is **the quality adjustment factor for product  $n$** .
- What is new is the assumption that **the quality adjustment factors are functions of a vector of  $K$  characteristics of the products**.
- Thus it is assumed that product  $n$  has the **vector of characteristics  $z^n \equiv [z_{n1}, z_{n2}, \dots, z_{nK}]$**  for  $n = 1, \dots, N$ .

## 10. The Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- The new assumption in this section is that the **quality adjustment factors**  $\alpha_n$  are **functions of the vector of characteristics**  $z^n$  for each product and the same function,  $f(z)$  can be used for each quality adjustment factor; i.e., we have the following assumptions:

$$(117) \alpha_n \equiv f(z^n) = f(z_{n1}, z_{n2}, \dots, z_{nK}) ; n = 1, \dots, N.$$

- Thus each product has its own unique mix of characteristics but **the same function  $f$  can be used to determine the relative utility to purchasers of the products.**
- Define the period  $t$  quantity vector as  $q^t = [q_{t1}, \dots, q_{tN}]$  for  $t = 1, \dots, T$ . If product  $n$  is missing in period  $t$ , then define  $q_{tn} \equiv 0$ .
- Using the above assumptions, **the aggregate quantity level  $Q^t$  for period  $t$  is defined as:**

$$(118) Q^t \equiv \alpha \cdot q^t = \sum_{n=1}^N f(z^n) q_{tn} ; t = 1, \dots, T.$$

## 10. The Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- Using our assumption of (exact) utility maximizing behavior with the linear utility function defined by (118), equations (10) become the following equations:

$$(119) \quad p_{tn} = \pi_t f(z^n) ; \quad t = 1, \dots, T; n \in S(t).$$

- The assumption of approximate utility maximizing behavior is more realistic so error terms need to be appended to equations (119).
- We also need to choose a functional form for the *quality adjustment (or hedonic) function*  $f(z)$ .
- Consider the following functional form for the **logarithm of the hedonic valuation function**:

$$(121) \quad \ln f(z^n) = \gamma_0 + \sum_{k=1}^K \gamma_k \ln z_{nk} = \ln \alpha_n \equiv \beta_n ; \quad n = 1, \dots, N.$$

- Now take logarithms of both sides of equations (119) and add error terms  $e_{tn}$  to the resulting equations.

## 10. The Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- Using equations (121), we obtain the following system of **estimating equations**:

$$(122) \ln p_{tn} = \rho_t + \gamma_0 + \sum_{k=1}^K \gamma_k \ln z_{nk} + e_{tn} ; \quad t = 1, \dots, T; n \in S(t)$$

- where as usual, we have defined  $\rho_t \equiv \ln \pi_t$  for  $t = 1, \dots, T$ .
- Equations (122) are the **usual log price hedonic regression equations**.
- Thus we have imbedded this model as a special case of our general class of utility maximization models which led to equations (10).
- Note that our derivation of this model rests on the **assumption of approximate utility maximizing behavior** on the part of purchasers of the  $N$  products.
- Note also that our underlying economic model assumes that the  $N$  products are **perfect substitutes** once they have been quality adjusted where the quality adjustment factors are defined by (117);  $\alpha_n \equiv f(z^n) = f(z_{n1}, z_{n2}, \dots, z_{nK})$ ;  $n = 1, \dots, N$ .

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- In previous sections, we noted that **weighting prices by their economic importance was generally recommended** (but not necessary if the fit of the corresponding hedonic regression was good).
- The same conclusion applies in the present context. Thus if quantity information is available (in addition to price and product characteristic information), then it is preferable to generate  $\rho$  and  $\gamma$  estimates by solving the following ***weighted least squares minimization problem***:

$$(130) \min_{\rho, \gamma} \sum_{t=1}^T \sum_{n \in S(t)} s_{tn} [\ln p_{tn} - \rho_t - \gamma_0 - \sum_{k=1}^K \gamma_k \ln z_{nk}]^2$$

- where the expenditure or sales shares  $s_{tn}$  are defined as  $s_{tn} \equiv p_{tn} q_{tn} / \sum_{i \in S(t)} p_{ti} q_{ti}$  for  $t = 1, \dots, T$  and  $n \in S(t)$ .
- A solution  $\rho, \gamma$  to the minimization problem (130) will satisfy the **following first order conditions**:

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

$$(131) \sum_{n \in S(t)} s_{tn} [\ln p_{tn} - \rho_t - \gamma_0 - \sum_{k=1}^K \gamma_k \ln z_{nk}] = 0 ; \quad t = 1, \dots, T;$$

$$(132) \sum_{t=1}^T \sum_{n \in S(t)} s_{tn} [\ln p_{tn} - \rho_t - \gamma_0 - \sum_{k=1}^K \gamma_k \ln z_{nk}] = 0 ;$$

$$(133) \sum_{t=1}^T \sum_{n \in S(t)} s_{tn} [\ln p_{tn} - \rho_t - \gamma_0 - \sum_{k=1}^K \gamma_k \ln z_{nk}] \ln z_{nk} = 0 ;$$

$k = 1, \dots, K.$

- Equations (131)-(133) are  $T+1+K$  equations in the  $T+1+K$  unknown parameters in the vectors  $\rho$  and  $\gamma$ .
- However, solutions to these equations are not unique; if  $\rho_t$  for  $t = 1, \dots, T$  and  $\gamma_k$  for  $k = 0, 1, \dots, K$  is a solution to (131)-(133), then  $\rho_t + \lambda$  for  $t = 1, \dots, T$ ,  $\gamma_0 - \lambda$  and  $\gamma_k$  for  $k = 1, \dots, K$  is also a solution for any number  $\lambda$ .
- Thus a **normalization on these parameters** is required for a unique solution to (131)-(133).
- Choose the normalization  $\rho_1^* = 0$  which is equivalent to  $\pi_1^* = 1$ .

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- Thus set  $\rho_1^* = 0$  in equations (131)-(133), drop the first equation in equations (131) and solve the remaining T+K equations for  $\rho_2^*, \dots, \rho_T^*$  and  $\gamma_0^*, \gamma_1^*, \dots, \gamma_K^*$ .
- Once these parameters have been determined, the estimated  $\beta_n^* \equiv \ln \alpha_n^*$  can be defined as  $\beta_n^* \equiv \gamma_0^* + \sum_{k=1}^K \gamma_k^* \ln z_{nk}$  for  $n = 1, \dots, N$ .
- Once the  $\beta_n^*$  have been defined, the corresponding quality adjustment factors are defined as  $\alpha_n^* \equiv \exp[\beta_n^*] > 0$  for  $n = 1, \dots, N$ .
- Using equations (131) evaluated at  $\rho^*$  and  $\gamma^*$ , we see that  $\pi_t^* \equiv \exp[\rho_t^*]$  is equal to the following expression:

$$(134) \pi_t^* = \exp[\sum_{n \in S(t)} s_{tn} \ln(p_{tn}/\alpha_n^*)] ; \quad t = 1, \dots, T$$

- with  $\pi_1^* \equiv 1$ . Thus the **period t estimated price level**,  $\pi_t^*$ , is an **expenditure share weighted geometric mean of the quality adjusted period t prices**,  $p_{tn}/\alpha_n^*$ , for the products n that are **present in period t**.

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- Once the  $\pi_t^*$  have been calculated, the **price index** between periods  $t$  and  $\tau$  is defined as  $\pi_t^*/\pi_\tau^*$  for  $1 \leq t, \tau \leq T$ .
- If quantity data are available, then we have the usual two methods for constructing period by period price and quantity levels,  $P^t$  and  $Q^t$  for  $t = 1, \dots, T$ ; see (110) and (111) above.
- $\pi_t^*/\pi_\tau^*$  is a share weighted geometric mean of the quality adjusted period  $t$  prices,  $p_{tn}/\alpha_n^*$ , for the products  $n$  that are present in period  $t$  with weights  $s_{tn}$  in the numerator divided by the share weighted geometric mean of the quality adjusted period  $\tau$  prices,  $p_{\tau n}/\alpha_n^*$ , for the products  $n$  that are present in period  $\tau$  with weights  $s_{\tau n}$  in the denominator.
- Thus **economic importance counts in the present model**.
- In the weighted time product dummy model, a single observation of a model did not affect the price levels. The weighted hedonic model with characteristics information does not have this unsatisfactory property. Every observation counts.

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

- Once the  $\rho^* \equiv [\rho_1^*, \rho_2^*, \dots, \rho_T^*]$  and  $\gamma^* \equiv [\gamma_0^*, \gamma_1^*, \dots, \gamma_K^*]$  solution to (130) has been determined (with  $\rho_1^* = 1$ ), we can use equations (122) to determine the **estimated residuals**  $e_{tn}^*$  for the model defined by (130).

- Thus we have the following equations:

$$\begin{aligned}
 (135) \quad e_{tn}^* &\equiv \ln p_{tn} - \rho_t^* - \gamma_0^* - \sum_{k=1}^K \gamma_k^* \ln z_{nk} ; & t = 1, \dots, T; n \in S(t) \\
 &= \ln p_{tn} - \rho_t^* - \beta_n^* & \text{using definitions (127)} \\
 &= \ln(p_{tn}/\pi_t^*) - \beta_n^* & \text{since } \rho_t^* \equiv \ln \pi_t^*.
 \end{aligned}$$

- Using definitions (135), it can be seen that the  $\beta_n^*$  satisfy the following equations:

$$(136) \quad \beta_n^* = \ln(p_{tn}/\pi_t^*) - e_{tn}^* ; \quad n = 1, \dots, N; t \in S^*(n).$$

- For each  $n$ , consider the following **share weighted average** of the  $\beta_n^*$  that appear in equations (136):

## 10. The Weighted Time Dummy Hedonic Regression Model with Characteristics Information (cont)

$$(137) \sum_{t \in S^*(n)} s_{tn} \beta_n^* = \sum_{t \in S^*(n)} s_{tn} [\ln(p_{tn}/\pi_t^*) - e_{tn}^*] ; \quad n = 1, \dots, N$$

$$\approx \sum_{t \in S^*(n)} s_{tn} \ln(p_{tn}/\pi_t^*)$$

- since the minimization problem defined by (130) will make the squared errors  $(e_{tn}^*)^2$  small within the constraints of the hedonic model.
  - Thus we have the following approximation for the  $\beta_n^*$ :
- $$(138) \beta_n^* \approx [\sum_{t \in S^*(n)} s_{tn} \ln(p_{tn}/\pi_t^*)] / \sum_{t \in S^*(n)} s_{tn} ; \quad n = 1, \dots, N.$$
- These equations provide **approximate counterparts** to equations (114) which were exact for the weighted time product dummy model discussed in section 9 above.
  - Thus **the logarithm of the product n quality adjustment factor,  $\beta_n^*$ , is approximately equal to a share weighted average of the logarithms of the inflation adjusted prices  $p_{tn}/\pi_t^*$  for product n over the periods t when this product was sold (or purchased) on the marketplace.**

## The de Haan and Krsinich Unit Value Problem

- Equations (136) and the equations  $\beta_n^* \equiv \ln \alpha_n^*$  for  $n = 1, \dots, N$  can be used to establish the following equations:

$$(139) \quad p_{tn} = \alpha_n^* \pi_t^* \exp[e_{tn}^*]; \quad t = 1, \dots, T.$$

- Suppose that the underlying hedonic model holds exactly so that each error term  $e_{tn}^*$  is equal to 0.
- Suppose further that all of the products are *perfect substitutes* so that the following equations hold:

$$(140) \quad \alpha_1 = \alpha_2 = \dots = \alpha_N.$$

- Then the estimated  $\alpha_n^*$  will equal  $\alpha_1^*$  for  $n = 1, \dots, N$  as well.
- Since the  $e_{tn}^* = 0$ ,  $\exp[e_{tn}^*] = 1$  for  $t = 1, \dots, T$ ;  $n \in S(t)$ .
- Substitute these relationships into equations (139). Now multiply both sides of equation  $tn$  in equations (139) by  $q_{tn}$  for  $t = 1, \dots, T$ ;  $n \in S(t)$ .
- We obtain the following system of equations after a certain amount of summation within each period:

## The de Haan and Krsinich Unit Value Problem (cont)

- Now take ratios of equations (141) for  $t = 1$  and a general  $t$ . After a bit of rearrangement, we obtain the following expression for the **price index** between periods 1 and  $t$ :

$$(142) \pi_t^*/\pi_1^* = \{\sum_{n \in S(t)} p_{tn} q_{tn} / \sum_{n \in S(t)} q_{tn}\} / \{\sum_{n \in S(1)} p_{1n} q_{1n} / \sum_{n \in S(1)} q_{1n}\};$$
$$t = 1, \dots, T.$$

- The right hand side of (142) for period  $t$  can be recognized as the **unit value price index** between periods 1 and  $t$ .
- The above algebra helps to resolve an **index number discontinuity problem** recognized by de Haan and Krsinich (2018; 760).
- These authors noted that the weighted geometric mean representation for  $\pi_t^* = \exp[\sum_{n \in S(t)} s_{tn} \ln(p_{tn}/\alpha_n^*)]$  (recall equations (134)) did not seem to collapse down to a unit value index if all of the estimated  $\alpha_n^*$  were equal which is disconcerting because if the products are perfect substitutes (without quality adjustment), then the appropriate index should collapse down to a unit value index (because each additional unit of any product gives the purchaser the same utility).

## The de Haan and Krsinich Unit Value Problem (cont)

- **However, if the products are perfect substitutes and markets are functioning properly, the price of every product in the group under consideration should be the same in each period.**
- **Under these conditions, the estimated  $\alpha_n^*$  will all be equal and equations (139) will become  $p_{tn} = \alpha_1^* \pi_t^*$  and equations (142) will hold.**
- **Thus under these conditions, there is no discontinuity problem; i.e., hedonic regression price index between periods t and 1 will collapse down to the unit value price index between periods t and 1.**
- **Thus the de Haan and Krsinich Unit Value Problem is resolved in a satisfactory manner.**

## The Remainder of the Paper

- We have shown that **carry forward pricing** is a special case of our general hedonic regression framework.
- **Superlative indexes** are also a special case of our framework that centers around equations (10).
- **Hedonic regressions** also fit into our framework.
- In the final sections of the paper, we look at **Hicksian reservation price methods** for quality adjustment. These methods involve some sort of econometric estimation.
- **Feenstra's reservation price methodology** only involves the estimation of an elasticity of substitution and once this estimate is in hand, his method is very easy to implement. However, since CES preferences can only provide a first order approximation to arbitrary preferences, Feenstra's method is likely to be biased: it will tend to overstate the benefits of new products.
- **Diewert and Feenstra** have developed a more general Hicksian reservation price methodology but it is too difficult to implement.

# Conclusion

- Using the theoretical framework explained in section 2 and applying it to hedonic regressions in section 5 (when price and quantity data are available) shows that the **hedonic regression approach generates two distinct estimates for the resulting price and quantity levels generated by the regression** (unless the regression fits the data perfectly, in which case the two methods generate identical estimates). Thus statistical agencies will have to choose between these two alternative index number estimates.
- **The use of weights that reflect economic importance is recommended** when running hedonic regressions; see the summary of the work of de Haan and Krsinich (2018) in section 11.
- In the two period context, section 7 shows how the **use of weights can transform** the problematic price index that results from an unweighted time product dummy hedonic regression into a superlative bilateral price index.

## Conclusion (cont)

- **The usefulness of the adjacent period weighted time product dummy hedonic regressions studied in section 8 is questionable; i.e., it may be preferable to use the carry forward and backward technique explained in section 4 in the bilateral case.**
- **Section 12 deals with hedonic regressions in the context of taste change. The results in this section indicate that the apparent increased flexibility offered by running separate hedonic regressions for each period is tempered by the need to average the results of the separate regressions in order to obtain proper price indexes. In the end, the use of a time dummy approach is recommended if the number of parameters is large relative to the number of observed prices.**
- **Hedonic regression models viewed from the Hicksian approach to the treatment of new products have a fundamental problem: the underlying economic model assumes that the products are perfect substitutes after the implied quality adjustment. This is not a problem if in fact, the quality adjusted products are close to being perfect substitutes but it can be a problem if this is not the case.**

## Conclusion (concluded)

- The **CES methodology** for accounting for the benefits of new products due to Feenstra explained in section 13 can work well if the elasticity of substitution between the products under consideration is high. If it is not high, the method will tend to lead to price indexes which have a **downward bias**.
- The **econometric method for dealing with new and disappearing products in the context of the Hicksian reservation price methodology** avoids the problems associated with the Feenstra methodology but at the cost of a great deal of econometric complexity. A robust simplified version of this methodology is required before it can be applied by statistical agencies on a routine basis.
- This paper has taken an economic approach to the problem of quality adjustment that is based on the basic model of household behavior explained in section 2. **This economic model is not without its problems but it does lead to a unified approach to the treatment of quality change from an economic perspective.**